

ROB521: Assignment 3

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April 6, 2023

1 Occupancy Map Algorithm

The mapped environment represents the ground truth from Assignment 2 and improves on the layout of the labyrinth and objects located in it. A balance between the factors α and β needs to be found when populating the map in order to obtain reasonably uniform contour of the wall - we are converting Cartesian coordinates to discrete grid coordinates and rounding error will cause cavities to appear along the walls, past the walls or within the walls (more noticeable at door passages).

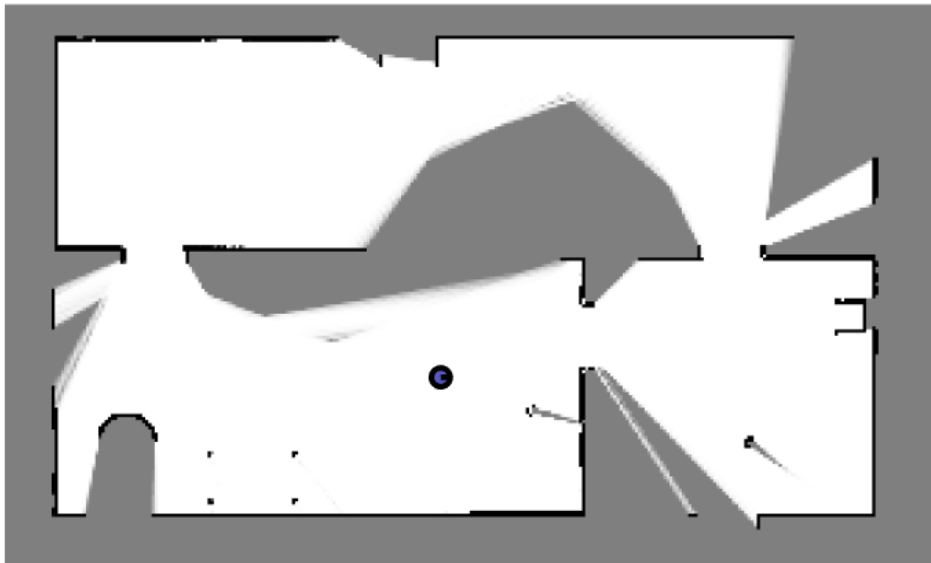


Figure 1: Mapping of the occupancy grid.

2 Particle Filter Localization

The results do not match the reference output exactly as we can observe a spike in the particle filter error when entering room three, which can be accounted to large input data ($\omega > 0.1 \text{ rad/s}$) in that particular sector coupled with missing data from the original map on the right side of the room. However, the particle filter manages to recover quickly and maintain lower error than the dead-reckoning approach - another possible fix for this would be in selecting a different distance metric when updating the weights, or more robust computations in the expected laser reading.

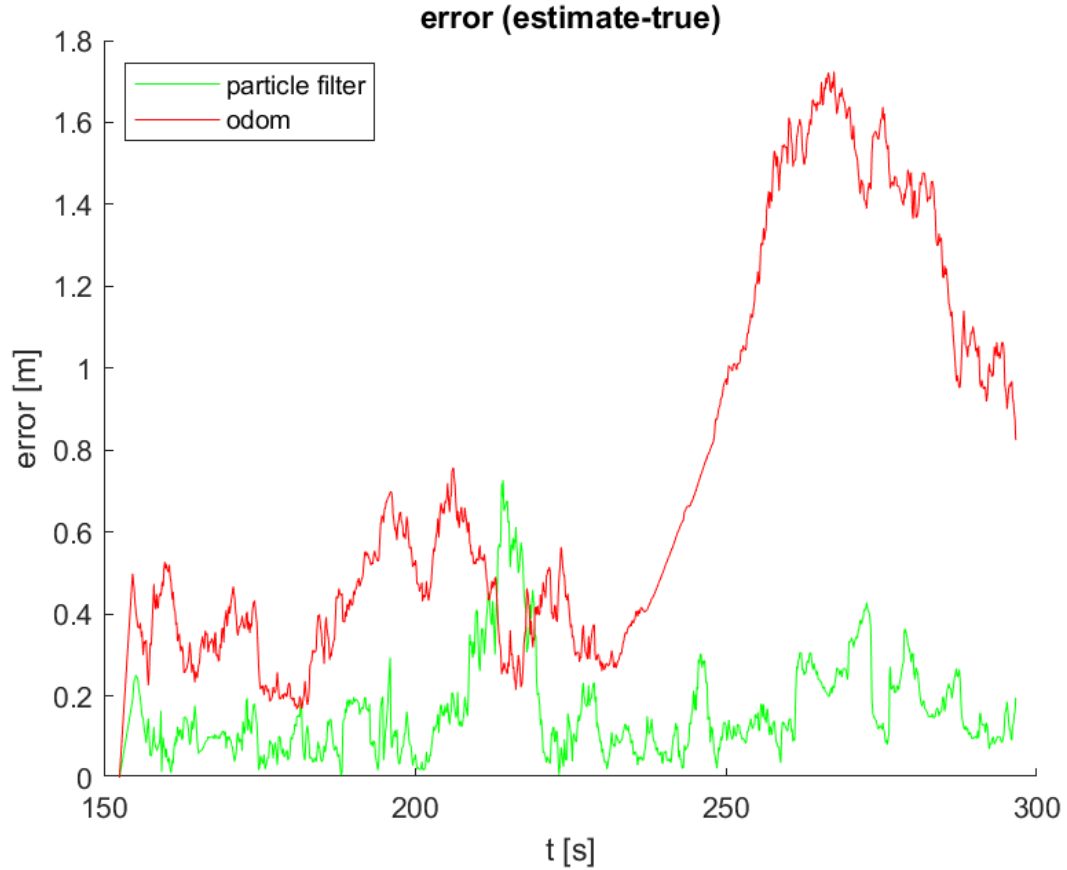


Figure 2: Dead-reckoning v. particle filter localization.

3 Source Code

3.1 Part I: Occupancy Mapping Algorithm

```
1 % =====
2 % ass3_q1.m
3 % =====
4 %
5 % This assignment will introduce you to the idea of first building an
6 % occupancy grid then using that grid to estimate a robot's motion using a
7 % particle filter.
8 %
9 % There are two questions to complete (5 marks each):
10 %
11 %     Question 1: code occupancy mapping algorithm
12 %     Question 2: see ass3_q2.m
13 %
14 % Fill in the required sections of this script with your code, run it to
15 % generate the requested plot/movie, then paste the plots into a short report
16 % that includes a few comments about what you've observed. Append your
17 % version of this script to the report. Hand in the report as a PDF file
18 % and the two resulting AVI files from Questions 1 and 2.
19 %
20 % requires: basic Matlab, 'gazebo.mat'
21 %
22 % T D Barfoot, January 2016
23 %
24 clear all; close all; clc;
25
26 % set random seed for repeatability
27 rng(1);
28
29 % =====
30 % load the dataset from file
31 % =====
32 %
33 %     ground truth poses: t_true x_true y_true theta_true
34 %     odometry measurements: t_odom v_odom omega_odom
35 %     laser scans: t_laser y_laser
36 %     laser range limits: r_min_laser r_max_laser
37 %     laser angle limits: phi_min_laser phi_max_laser
38 %
39 load gazebo.mat;
40
41 % =====
42 % Question 1: build an occupancy grid map
43 % =====
44 %
45 % Write an occupancy grid mapping algorithm that builds the map from the
46 % perfect ground-truth localization. Some of the setup is done for you
47 % below. The resulting map should look like "ass2_q1_soln.png". You can
48 % watch the movie "ass2_q1_soln.mp4" to see what the entire mapping process
49 % should look like. At the end you will save your occupancy grid map to
50 % the file "occmap.mat" for use in Question 2 of this assignment.
51
52 % allocate a big 2D array for the occupancy grid
53 ogres = 0.05; % resolution of occ grid
54 ogxmin = -7; % minimum x value
55 ogxmax = 8; % maximum x value
56 ogymmin = -3; % minimum y value
57 ogymax = 6; % maximum y value
58 ognx = (ogxmax-ogxmin)/ogres; % number of cells in x direction
59 ogny = (ogymax-ogymmin)/ogres; % number of cells in y direction
60 oglo = zeros(ogny,ognx); % occupancy grid in log-odds format
61 ogp = zeros(ogny,ognx); % occupancy grid in probability format
62
63 % precalculate some quantities
64 numodom = size(t_odom,1);
65 npoints = size(y_laser,2);
66 angles = linspace(phi_min_laser, phi_max_laser,npoints);
67 dx = ogres*cos(angles);
68 dy = ogres*sin(angles);
69
```

```

70 % interpolate the noise-free ground-truth at the laser timestamps
71 t_interp = linspace(t_true(1),t_true(numodom),numodom);
72 x_interp = interp1(t_interp,x_true,t_laser);
73 y_interp = interp1(t_interp,y_true,t_laser);
74 theta_interp = interp1(t_interp,theta_true,t_laser);
75 omega_interp = interp1(t_interp,omega_odom,t_laser);
76
77 % set up the plotting/movie recording
78 vid = VideoWriter('ass2_q1.avi');
79 open(vid);
80 figure(1);
81 clf;
82 pcolor(ogp);
83 colormap(1-gray);
84 shading('flat');
85 axis equal;
86 axis off;
87 M = getframe;
88 writeVideo(vid,M);
89
90 % precalculate some quantities
91 cos_angles = cos(angles);
92 sin_angles = sin(angles);
93 offset = [0.1; 0];
94 map0 = [ogxmin; ogymin];
95
96 alpha = 2.5;
97 beta = 0.15;
98
99 % loop over laser scans (every fifth)
100 for i=1:5:size(t_laser,1)
101
102     % -----insert your occupancy grid mapping algorithm here-----
103
104     if abs(omega_interp(i)) < 0.1
105
106         % extract robot pose in 2D space
107         thetaR = theta_interp(i);
108         ptR = [x_interp(i); y_interp(i)];
109         R_i = [cos(-thetaR), sin(-thetaR);
110               -sin(-thetaR), cos(-thetaR)];
111
112         % transform all laser scans to robot's local frame in one go
113         RLaser_i = y_laser(i, :);
114         QLaser_i = [RLaser_i.*cos_angles; RLaser_i.*sin_angles];
115
116         QGlobal = R_i*(QLaser_i - offset) + ptR - map0;
117
118         % parse entries 1-by-1 to update the log likelihoods
119         for j = 1 : npoints
120
121             % skip NaN data entries
122             if isnan(RLaser_i(j))
123                 continue
124             end
125
126             % check ranges to accumulate log likely hoods
127             for r = r_min_laser : ogres : r_max_laser
128
129                 if r < RLaser_i(j)
130
131                     % transform range in global frame
132                     rG = ptR + R_i*(r*[cos_angles(j); sin_angles(j)] - offset) - map0;
133
134                     % reward empty region
135                     oglo(floor(rG(2)/ogres), floor(rG(1)/ogres)) = oglo(floor(rG(2)/
ogres), floor(rG(1)/ogres)) - beta;
136
137                 end
138
139             end
140
141         % penalize occupied region

```

```

142         oglo(floor(QGlobal(2,j)/ogres), floor(QGlobal(1,j)/ogres)) = oglo(floor(
143         QGlobal(2, j)/ogres), floor(QGlobal(1,j)/ogres)) + alpha;
144
145     end
146
147     % convert log probabilities to probabilities
148     for ogpX = 1:size(ogp, 1)
149         for ogpY = 1:size(ogp, 2)
150             ogp(ogpX, ogpY) = exp(oglo(ogpX, ogpY)) / (1 + exp(oglo(ogpX, ogpY)));
151         end
152     end
153
154     % -----end of your occupancy grid mapping algorithm-----
155
156     % draw the map
157     clf;
158     pcolor(ogp);
159     colormap(1-gray);
160     shading('flat');
161     axis equal;
162     axis off;
163
164     % draw the robot
165     hold on;
166     x = (x_interp(i)-ogxmin)/ogres;
167     y = (y_interp(i)-ogymin)/ogres;
168     th = theta_interp(i);
169     r = 0.15/ogres;
170     set(rectangle('Position', [x-r y-r 2*r 2*r], 'Curvature', [1 1]), 'LineWidth', 2, '
171     FaceColor', [0.35 0.35 0.75]);
172     set(plot([x x+r*cos(th)], [y y+r*sin(th)], 'k-'), 'LineWidth', 2);
173
174     % save the video frame
175     M = getframe;
176     writeVideo(vid,M);
177
178     pause(0.1);
179 end
180
181 close(vid);
182 print -dpng ass2_q1.png
183
184 save occmap.mat ogres ogxmin ogxmax ogymmin ogymax ognx ogny oglo ogp;

```

3.2 Part II: Particle Filter Algorithm

```

1 % =====
2 % ass3_q2.m
3 % =====
4 %
5 % This assignment will introduce you to the idea of first building an
6 % occupancy grid then using that grid to estimate a robot's motion using a
7 % particle filter.
8 %
9 % There are three questions to complete (5 marks each):
10 %
11 %     Question 1: see ass3_q1.m
12 %     Question 2: code particle filter to localize from known map
13 %
14 % Fill in the required sections of this script with your code, run it to
15 % generate the requested plot/movie, then paste the plots into a short report
16 % that includes a few comments about what you've observed. Append your
17 % version of this script to the report. Hand in the report as a PDF file
18 % and the two resulting AVI files from Questions 1 and 2.
19 %
20 % requires: basic Matlab, 'gazebo.mat', 'occmap.mat'
21 %
22 % T D Barfoot, January 2016
23 %
24 clear all; close all; clc;
25

```

```

26 % set random seed for repeatability
27 rng(1);
28
29 % =====
30 % load the dataset from file
31 % =====
32 %
33 %     ground truth poses: t_true x_true y_true theta_true
34 %     odometry measurements: t_odom v_odom omega_odom
35 %     laser scans: t_laser y_laser
36 %     laser range limits: r_min_laser r_max_laser
37 %     laser angle limits: phi_min_laser phi_max_laser
38 %
39 load gazebo.mat;
40
41 % =====
42 % load the occupancy map from question 1 from file
43 % =====
44 % ogres: resolution of occ grid
45 % ogxmin: minimum x value
46 % ogxmax: maximum x value
47 % ogymmin: minimum y value
48 % ogymax: maximum y value
49 % ognx: number of cells in x direction
50 % ognx: number of cells in y direction
51 % oglo: occupancy grid in log-odds format
52 % ogp: occupancy grid in probability format
53 load occmap.mat;
54
55 % =====
56 % Question 2: localization from an occupancy grid map using particle filter
57 % =====
58 %
59 % Write a particle filter localization algorithm to localize from the laser
60 % rangefinder readings, wheel odometry, and the occupancy grid map you
61 % built in Question 1. We will only use two laser scan lines at the
62 % extreme left and right of the field of view, to demonstrate that the
63 % algorithm does not need a lot of information to localize fairly well. To
64 % make the problem harder, the below lines add noise to the wheel odometry
65 % and to the laser scans. You can watch the movie "ass2_q2_soln.mp4" to
66 % see what the results should look like. The plot "ass2_q2_soln.png" shows
67 % the errors in the estimates produced by wheel odometry alone and by the
68 % particle filter look like as compared to ground truth; we can see that
69 % the errors are much lower when we use the particle filter.
70
71 % interpolate the noise-free ground-truth at the laser timestamps
72 numodom = size(t_odom,1);
73 t_interp = linspace(t_true(1),t_true(numodom),numodom);
74 x_interp = interp1(t_interp,x_true,t_laser);
75 y_interp = interp1(t_interp,y_true,t_laser);
76 theta_interp = interp1(t_interp,theta_true,t_laser);
77 omega_interp = interp1(t_interp,omega_odom,t_laser);
78
79 % interpolate the wheel odometry at the laser timestamps and
80 % add noise to measurements (yes, on purpose to see effect)
81 v_interp = interp1(t_interp,v_odom,t_laser) + 0.2*randn(size(t_laser,1),1);
82 omega_interp = interp1(t_interp,omega_odom,t_laser) + 0.04*randn(size(t_laser,1),1);
83
84 % add noise to the laser range measurements (yes, on purpose to see effect)
85 % and precompute some quantities useful to the laser
86 y_laser = y_laser + 0.1*randn(size(y_laser));
87 npoints = size(y_laser,2);
88 angles = linspace(phi_min_laser, phi_max_laser,npoints);
89 dx = ogres*cos(angles);
90 dy = ogres*sin(angles);
91 y_laser_max = 5; % don't use laser measurements beyond this distance
92
93 % particle filter tuning parameters (yours may be different)
94 nparticles = 200; % number of particles
95 v_noise = 0.2; % noise on longitudinal speed for propagating particle
96 u_noise = 0.2; % noise on lateral speed for propagating particle
97 omega_noise = 0.04; % noise on rotational speed for propagating particle
98 laser_var = 0.5^2; % variance on laser range distribution

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99 w_gain = 10*sqrt( 2 * pi * laser_var );      % gain on particle weight
100
101 % generate an initial cloud of particles
102 x_particle = x_true(1) + 0.5*randn(nparticles,1);
103 y_particle = y_true(1) + 0.3*randn(nparticles,1);
104 theta_particle = theta_true(1) + 0.1*randn(nparticles,1);
105
106 % compute a wheel odometry only estimate for comparison to particle
107 % filter
108 x_odom_only = x_true(1);
109 y_odom_only = y_true(1);
110 theta_odom_only = theta_true(1);
111
112 % error variables for final error plots - set the errors to zero at the start
113 pf_err(1) = 0;
114 wo_err(1) = 0;
115
116 % set up the plotting/movie recording
117 vid = VideoWriter('ass2_q2.avi');
118 open(vid);
119 figure(2);
120 clf;
121 hold on;
122 pcolor(ogp);
123 set(plot( (x_particle-ogxmin)/ogres, (y_particle-ogymin)/ogres, 'g.' ),'MarkerSize',10,'
    Color',[0 0.6 0]);
124 set(plot( (x_odom_only-ogxmin)/ogres, (y_odom_only-ogymin)/ogres, 'r.' ),'MarkerSize'
    ,20);
125 x = (x_interp(1)-ogxmin)/ogres;
126 y = (y_interp(1)-ogymin)/ogres;
127 th = theta_interp(1);
128 r = 0.15/ogres;
129 set(rectangle( 'Position', [x-r y-r 2*r 2*r], 'Curvature', [1 1]),'LineWidth',2,'
    FaceColor',[0.35 0.35 0.75]);
130 set(plot([x x+r*cos(th)]', [y y+r*sin(th)]', 'k-'),'LineWidth',2);
131 set(plot( (mean(x_particle)-ogxmin)/ogres, (mean(y_particle)-ogymin)/ogres, 'g.' ),'
    MarkerSize',20);
132 colormap(1-gray);
133 shading('flat');
134 axis equal;
135 axis off;
136 M = getframe;
137 writeVideo(vid,M);
138
139 % loop over laser scans
140 for i=2:size(t_laser,1)
141
142     % update the wheel-odometry-only algorithm
143     dt = t_laser(i) - t_laser(i-1);
144     v = v_interp(i);
145     omega = omega_interp(i);
146     x_odom_only = x_odom_only + dt*v*cos( theta_odom_only );
147     y_odom_only = y_odom_only + dt*v*sin( theta_odom_only );
148     phi = theta_odom_only + dt*omega;
149     while phi > pi
150         phi = phi - 2*pi;
151     end
152     while phi < -pi
153         phi = phi + 2*pi;
154     end
155     theta_odom_only = phi;
156
157     % loop over the particles
158     for n=1:nparticles
159
160         % propagate the particle forward in time using wheel odometry
161         % (remember to add some unique noise to each particle so they
162         % spread out over time)
163         v = v_interp(i) + v_noise*randn(1);
164         u = u_noise*randn(1);
165         omega = omega_interp(i) + omega_noise*randn(1);
166         x_particle(n) = x_particle(n) + dt*(v*cos( theta_particle(n) ) - u*sin(
            theta_particle(n) ));

```

```

167     y_particle(n) = y_particle(n) + dt*(v*sin( theta_particle(n) ) + u*cos(
theta_particle(n) ));
168     phi = theta_particle(n) + dt*omega;
169     while phi > pi
170         phi = phi - 2*pi;
171     end
172     while phi < -pi
173         phi = phi + 2*pi;
174     end
175     theta_particle(n) = phi;
176
177     % pose of particle in initial frame
178     T = [cos(theta_particle(n)) -sin(theta_particle(n)) x_particle(n); ...
179         sin(theta_particle(n))  cos(theta_particle(n)) y_particle(n); ...
180         0                        0                        1];
181
182     % compute the weight for each particle using only 2 laser rays
183     % (right=beam 1 and left=beam 640)
184     w_particle(n) = 1.0;
185     for beam=1:2
186
187         % we will only use the first and last laser ray for
188         % localization
189         if beam==1 % rightmost beam
190             j = 1;
191         elseif beam==2 % leftmost beam
192             j = 640;
193         end
194
195         % -----insert your particle filter weight calculation here -----
196
197         % compute bearing angle in robot's frame of reference
198         phi_j_particle = angles(j);
199
200         % initialize variables to 0
201         xGrid = 0;
202         yGrid = 0;
203         R = 0;
204
205         % iterate through viable ranges towards the bearing angle
206         % direction
207         for r = r_min_laser : ogres : r_max_laser
208
209             % pose transformation to global coordinates and grid (x,y)
210             rParticle = r*[cos(phi_j_particle); sin(phi_j_particle)] - [0.1; 0];
211             rGlobal = T*[rParticle; 1];
212
213             xGrid = round((rGlobal(1) - ogxmin)/ogres);
214             yGrid = round((rGlobal(2) - ogymmin)/ogres);
215
216             % adjust values
217             if xGrid > ognx
218                 xGrid = ognx;
219             end
220
221             if yGrid > ogny
222                 yGrid = ogny;
223             end
224
225             if yGrid < 1
226                 yGrid = 1;
227             end
228
229             if xGrid < 1
230                 xGrid = 1;
231             end
232
233             % check the value of the occupancy map
234             if ogp(yGrid, xGrid) > 0.5
235                 R = r;
236                 break
237             end
238

```



```

239         end
240
241         % compute gain
242         w_particle(n) = w_gain*w_particle(n)*normpdf(y_laser(i,j),...
243             R, sqrt(laser_var));
244
245         % -----end of your particle filter weight calculation-----
246     end
247
248 end
249
250 % resample the particles using Madow systematic resampling
251 w_bounds = cumsum(w_particle)/sum(w_particle);
252 w_target = rand(1);
253 j = 1;
254 for n=1:nparticles
255     while w_bounds(j) < w_target
256         j = mod(j,nparticles) + 1;
257     end
258     x_particle_new(n) = x_particle(j);
259     y_particle_new(n) = y_particle(j);
260     theta_particle_new(n) = theta_particle(j);
261     w_target = w_target + 1/nparticles;
262     if w_target > 1
263         w_target = w_target - 1.0;
264         j = 1;
265     end
266 end
267 x_particle = x_particle_new;
268 y_particle = y_particle_new;
269 theta_particle = theta_particle_new;
270
271 % save the translational error for later plotting
272 pf_err(i) = sqrt( (mean(x_particle) - x_interp(i))^2 + (mean(y_particle) - y_interp(
273 i))^2 );
274 wo_err(i) = sqrt( (x_odom_only - x_interp(i))^2 + (y_odom_only - y_interp(i))^2 );
275
276 % plotting
277 figure(2);
278 clf;
279 hold on;
280 pcolor(ogp);
281 set(plot( (x_particle-ogxmin)/ogres, (y_particle-ogymin)/ogres, 'g.' ),'MarkerSize'
282 ,10,'Color',[0 0.6 0]);
283 set(plot( (x_odom_only-ogxmin)/ogres, (y_odom_only-ogymin)/ogres, 'r.' ),'MarkerSize'
284 ,20);
285 x = (x_interp(i)-ogxmin)/ogres;
286 y = (y_interp(i)-ogymin)/ogres;
287 th = theta_interp(i);
288 if ~isnan(y_laser(i,1)) & y_laser(i,1) <= y_laser_max
289     set(plot([x x+y_laser(i,1)/ogres*cos(th+angles(1))]', [y y+y_laser(i,1)/ogres*sin
290 (th+angles(1))]', 'm-'),'LineWidth',1);
291 end
292 if ~isnan(y_laser(i,640)) & y_laser(i,640) <= y_laser_max
293     set(plot([x x+y_laser(i,640)/ogres*cos(th+angles(640))]', [y y+y_laser(i,640)/
294 ogres*sin(th+angles(640))]', 'm-'),'LineWidth',1);
295 end
296 r = 0.15/ogres;
297 set(rectangle( 'Position', [x-r y-r 2*r 2*r], 'Curvature', [1 1]),'LineWidth',2,'
298 FaceColor',[0.35 0.35 0.75]);
299 set(plot([x x+r*cos(th)]', [y y+r*sin(th)]', 'k-'),'LineWidth',2);
300 set(plot( (mean(x_particle)-ogxmin)/ogres, (mean(y_particle)-ogymin)/ogres, 'g.' ),'
301 MarkerSize',20);
302 colormap(1-gray);
303 shading('flat');
304 axis equal;
305 axis off;
306
307 % save the video frame
308 M = getframe;
309 writeVideo(vid,M);
310
311 pause(0.005);

```

```

305
306 end
307
308 close(vid);
309
310 % final error plots
311 figure(3);
312 clf;
313 hold on;
314 plot( t_laser, pf_err, 'g-' );
315 plot( t_laser, wo_err, 'r-' );
316 xlabel('t [s]');
317 ylabel('error [m]');
318 legend('particle filter', 'odom', 'Location', 'NorthWest');
319 title('error (estimate-true)');
320 print -dpng ass2_q2.png

```